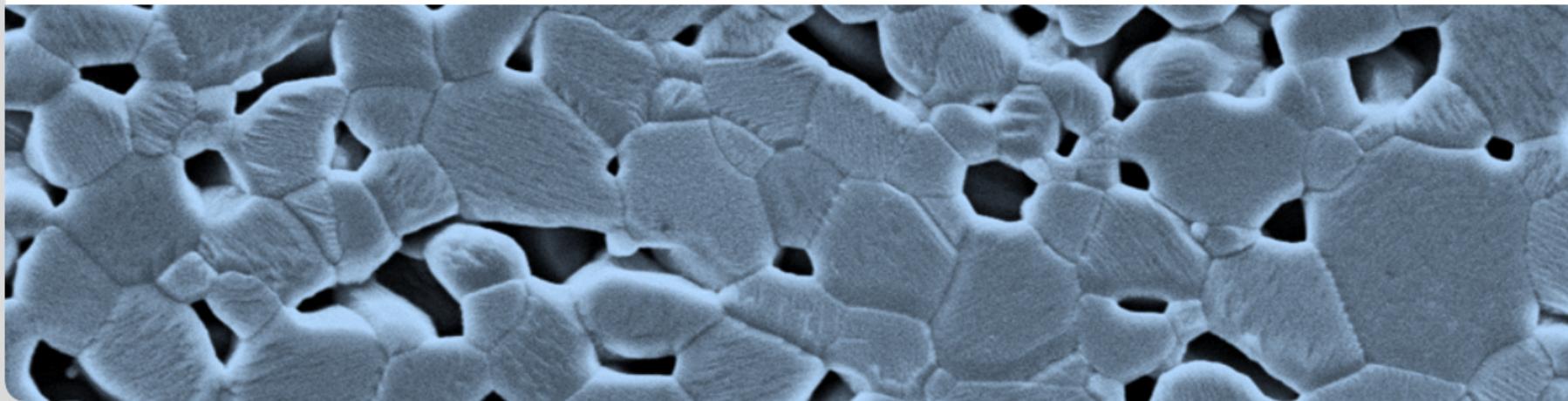


Extreme Scale Phase-Field Simulations of Sintering Processes

J. Hötzer, H. Hierl, M. Seiz, A. Reiter and B. Nestler

INSTITUTE OF APPLIED MATERIALS - COMPUTATIONAL MATERIALS SCIENCE

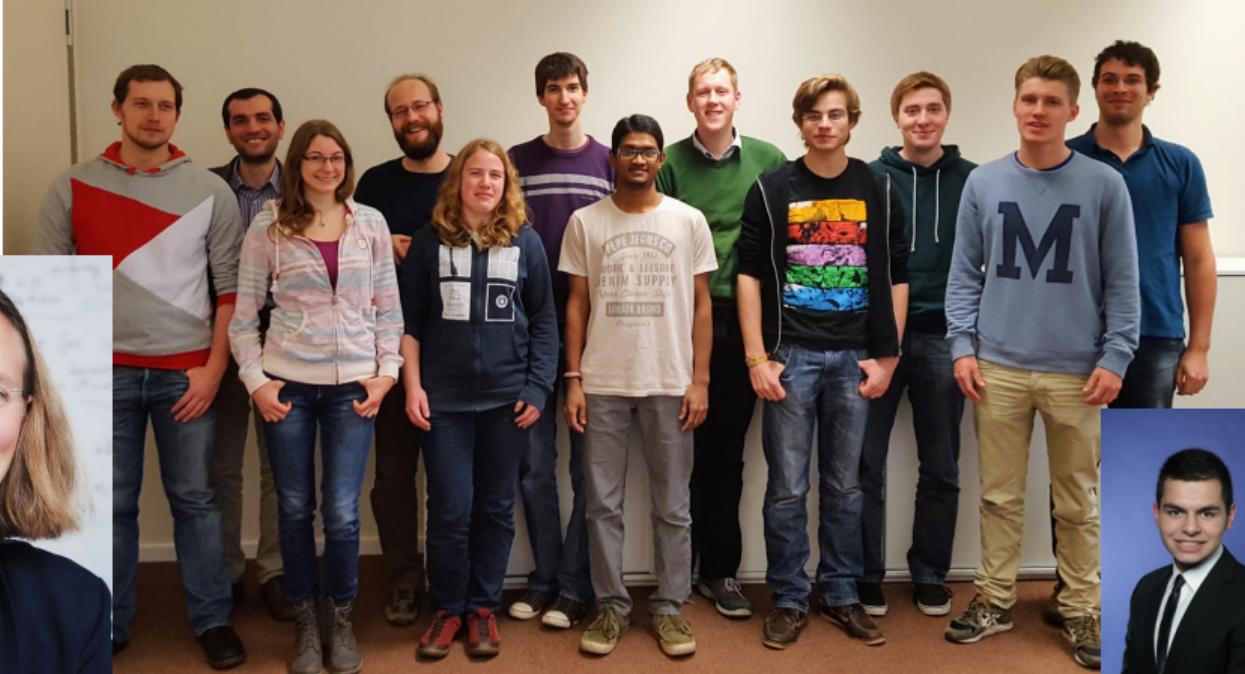




Part I: Institute

<http://zkm.de/en/event/2015/06/schlosslichtspiele>

The Group “High Performance Materials Computing and Data Science”

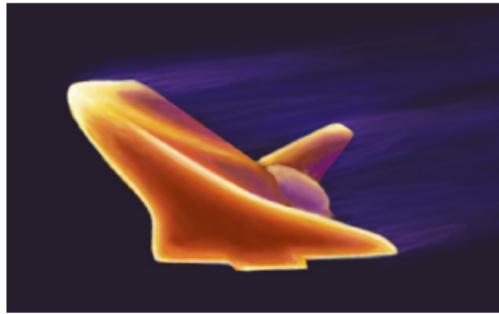




Part II: Motivation

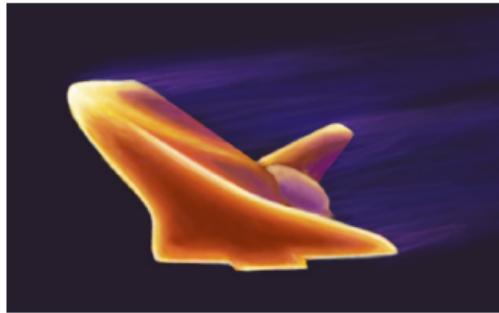
KERAMIK

What do these pictures have in common?



<https://upload.wikimedia.org/wikipedia/commons/6/6a/Sparkplug.jpg>

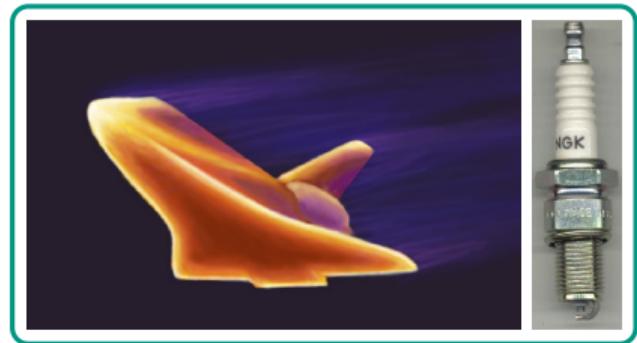
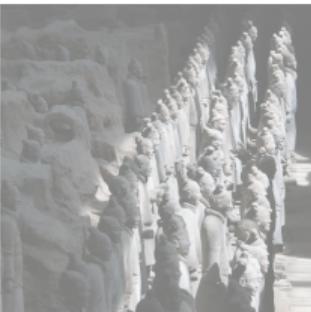
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all structures result from the same process:
sintering

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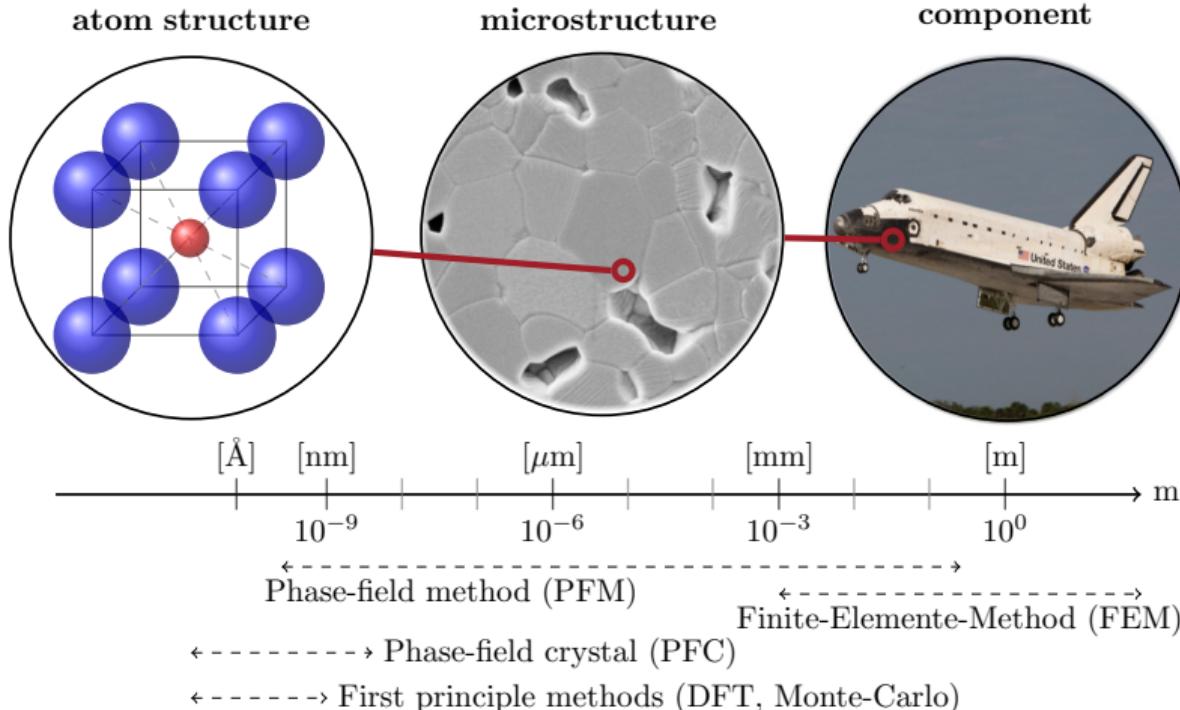


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high performance materials require defined properties

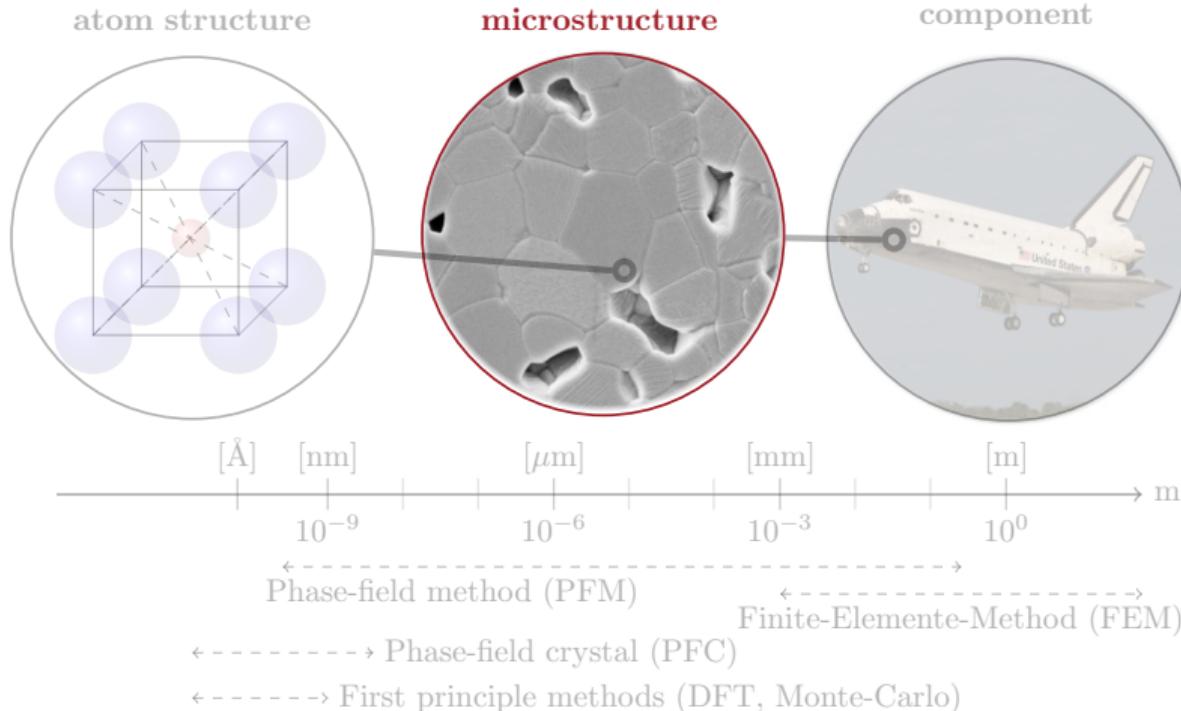
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Motivation - Material properties



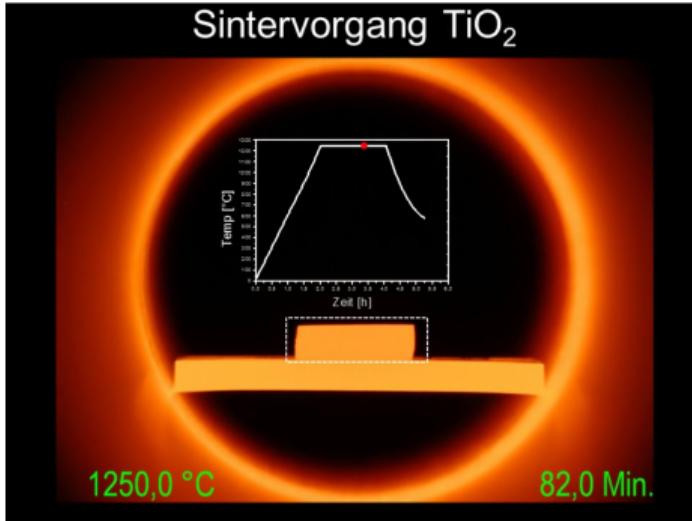
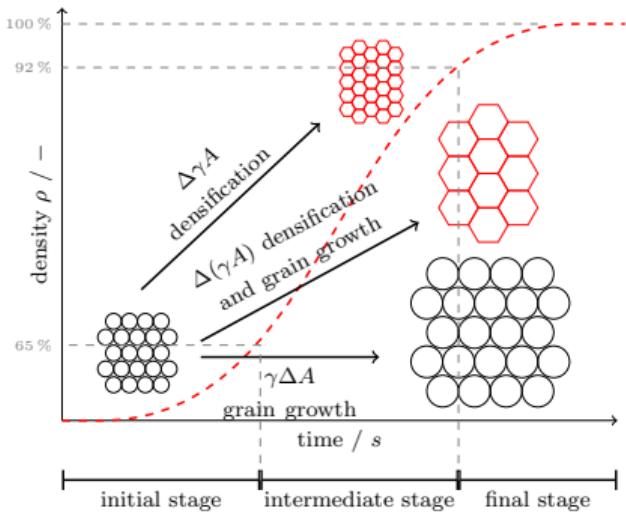
■ **chemical elements** and **microstructure** determine the **properties** of the component
⇒ controlling the **microstructure evolution** = controlling the **material properties**

Motivation - Material properties



■ **chemical elements** and **microstructure** determine the **properties** of the component
 ⇒ controlling the **microstructure evolution** = controlling the **material properties**

Sintering process



- three stages of sintering
 - initial: evolution of sintering necks
 - intermediate: **densification** of green body
 - final: **grain growth**

⇒ difficult to predict and control microstructure

Video: IAM-KWT, KIT

$$\Psi(\phi, \mu, T) = \int_{\Omega} (\varepsilon \tilde{a}(\phi, \nabla \phi) + \frac{1}{\varepsilon} \omega(\phi)) + \psi(\phi, \mu, T) d\Omega$$

$$\tau \varepsilon \frac{\partial \phi}{\partial t} = -\varepsilon \left(\frac{\partial a(\phi, \nabla \phi)}{\partial \phi_\alpha} + \nabla \cdot \frac{\partial a(\phi, \nabla \phi)}{\partial \nabla \phi_\alpha} \right) - \frac{1}{N, N} \frac{\partial \omega(\phi)}{\partial \phi_\alpha} - \frac{\partial \psi(\phi, \mu, T)}{\partial \phi_\alpha} + \lambda$$

$$h_\alpha(\phi) = \frac{\phi_\alpha^2}{\sum_{\beta=1}^N \phi_\beta^2}$$

$$\omega(\phi) = \begin{cases} \frac{16}{\pi^2} \sum_{\substack{\alpha, \beta=1 \\ (\alpha < \beta)}} \gamma_{\alpha\beta} \phi_\alpha \phi_\beta + \sum_{\substack{\alpha, \beta, \delta=1 \\ (\alpha < \beta < \delta)}} \gamma_{\alpha\beta\delta} \phi_\alpha \phi_\beta \phi_\delta, & \phi \in \Delta^{N-1} \\ \phi \notin \Delta^{N-1} & \text{otherwise} \end{cases}$$

$X_\alpha(T) = \tilde{F}_\alpha(T)$

$f_\alpha(\vec{c}, T) = \langle \vec{c}, \Xi_\alpha(T) \vec{c} \rangle + \langle \vec{c}, \Xi_\alpha(T) \vec{c} \rangle$

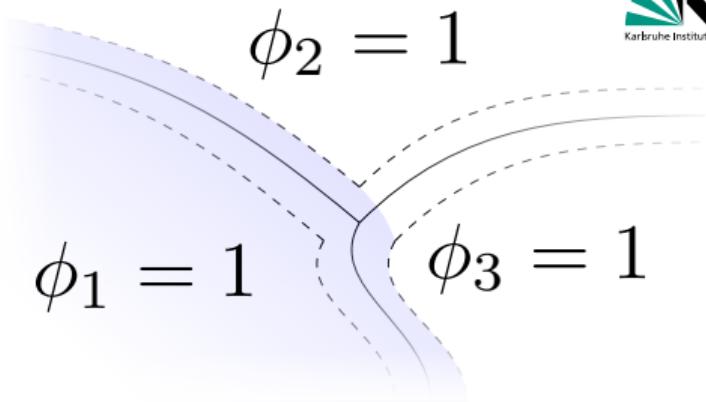
$$a(\phi, \nabla \phi) = \sum_{\substack{\alpha, \beta=1 \\ (\alpha < \beta)}}^{N, N} \gamma_{\alpha\beta} |q_{\alpha\beta}|^2 \quad S(\phi, \nabla \phi) = \int_{\Omega} \varepsilon \tilde{a}(\phi, \nabla \phi) + \frac{1}{\varepsilon} \tilde{\omega}(\phi) d\mathbf{x}$$

$$\frac{\partial \mu}{\partial t} = \left[\left(\frac{\partial \vec{c}}{\partial \mu} \right)_{T, \vec{\phi}} \right]^{-1} \left(\nabla \cdot (\mathbf{M}(\phi, T) \nabla \mu) - \left(\frac{\partial \vec{c}}{\partial \phi} \right)_{T, \mu} \frac{\partial \vec{\phi}}{\partial t} - \left(\frac{\partial \vec{c}}{\partial T} \right)_{\mu, \vec{\phi}} \frac{\partial T}{\partial t} \right)$$

$$J_{\text{at}} = \frac{\pi \varepsilon}{4} \sum_{\substack{\alpha=1 \\ (\alpha \neq \ell)}}^N \frac{h_\alpha(\vec{\phi}) h_\ell(\vec{\phi})}{\sqrt{\phi_\alpha \phi_\ell}} \frac{\partial \phi_\alpha}{\partial t} \left(\left\langle \frac{\nabla \phi_\alpha}{|\nabla \phi_\alpha|}, \frac{\nabla \phi_\ell}{|\nabla \phi_\ell|} \right\rangle \right) \left((\vec{c}^\ell(\mu, T) - \vec{c}^\alpha(\mu, T)) \otimes \frac{\nabla \phi_\alpha}{|\nabla \phi_\alpha|} \right)$$

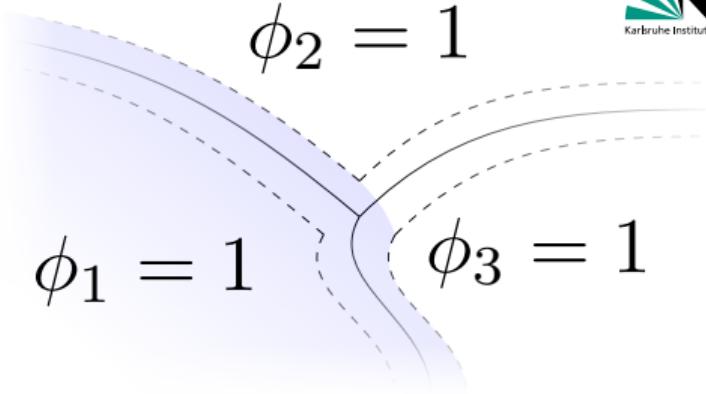
Phase-field method

- **minimization** of the **total energy** in a system under constraints
- **coupling** with different **physical quantities** (e.g. concentration, temperature, ...)
- **phase-field vector** $\phi = (\phi_1, \phi_2, \dots, \phi_N)^T$
- **order-parameter** ϕ_α describes the volume fraction of the phases with different properties, e.g. orientation, ...
- modeling of a defined **diffuse interface** between the phases
 - ⇒ for numerical resolution
 - ≈ 10 cells required
 - ⇒ high resolution of the structures required



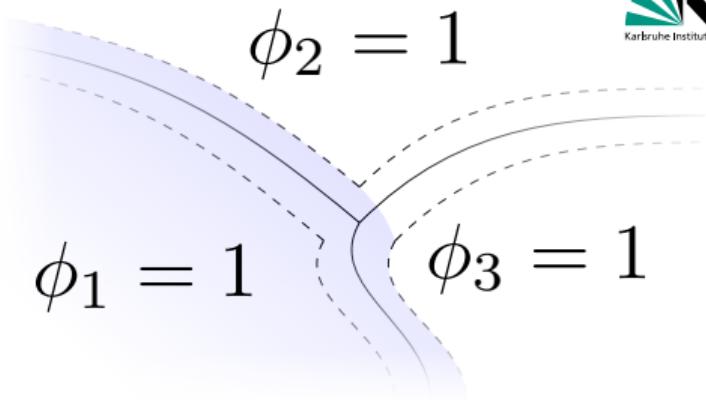
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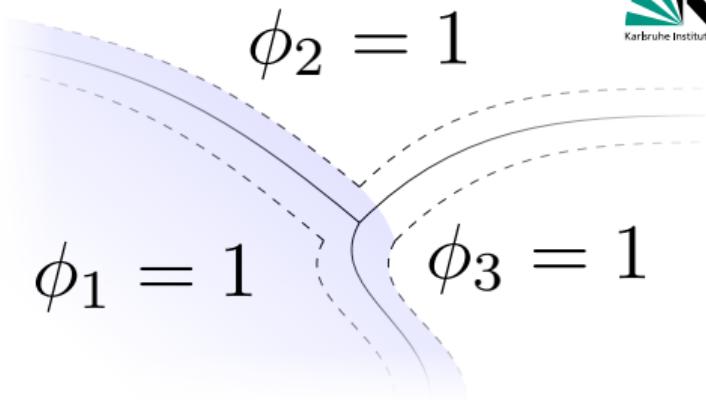
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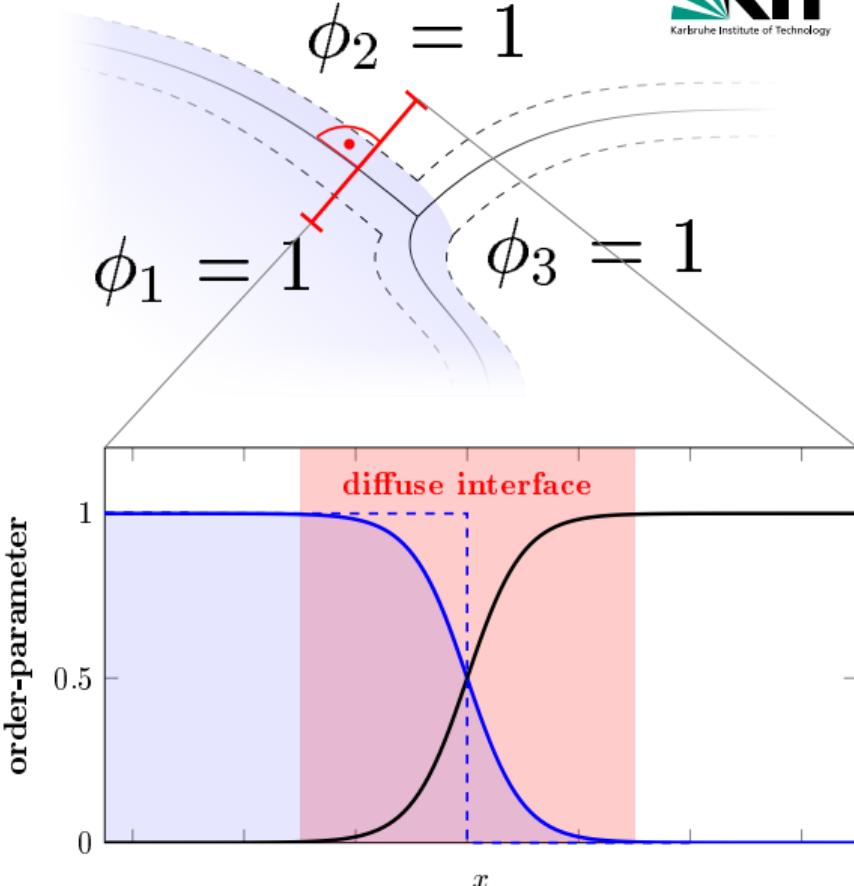
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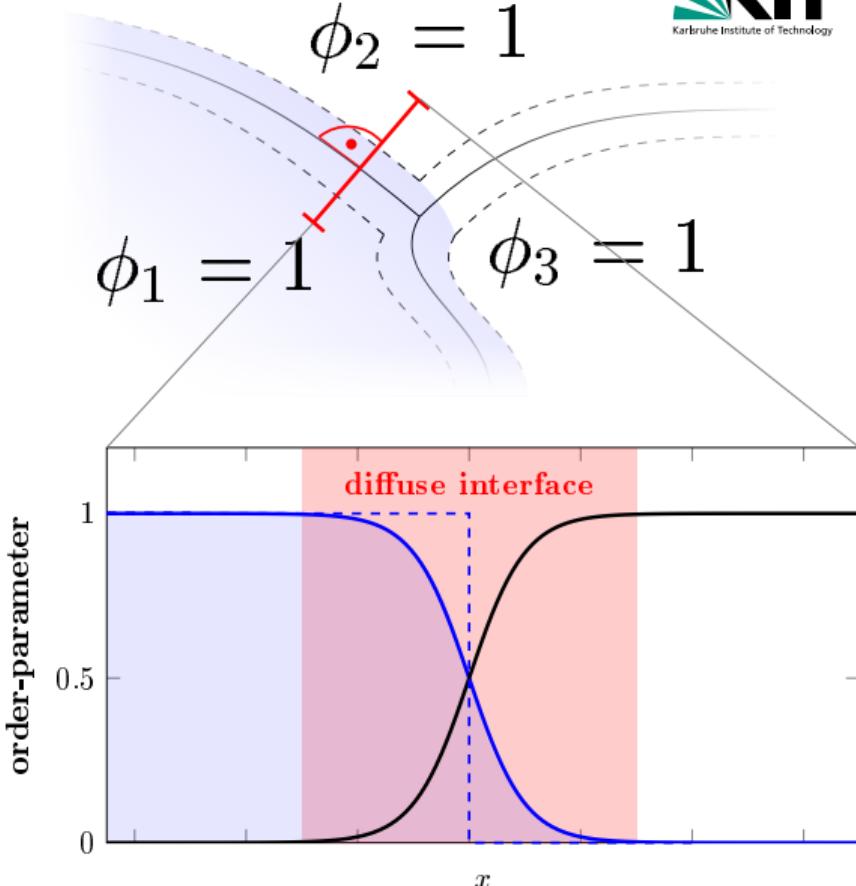
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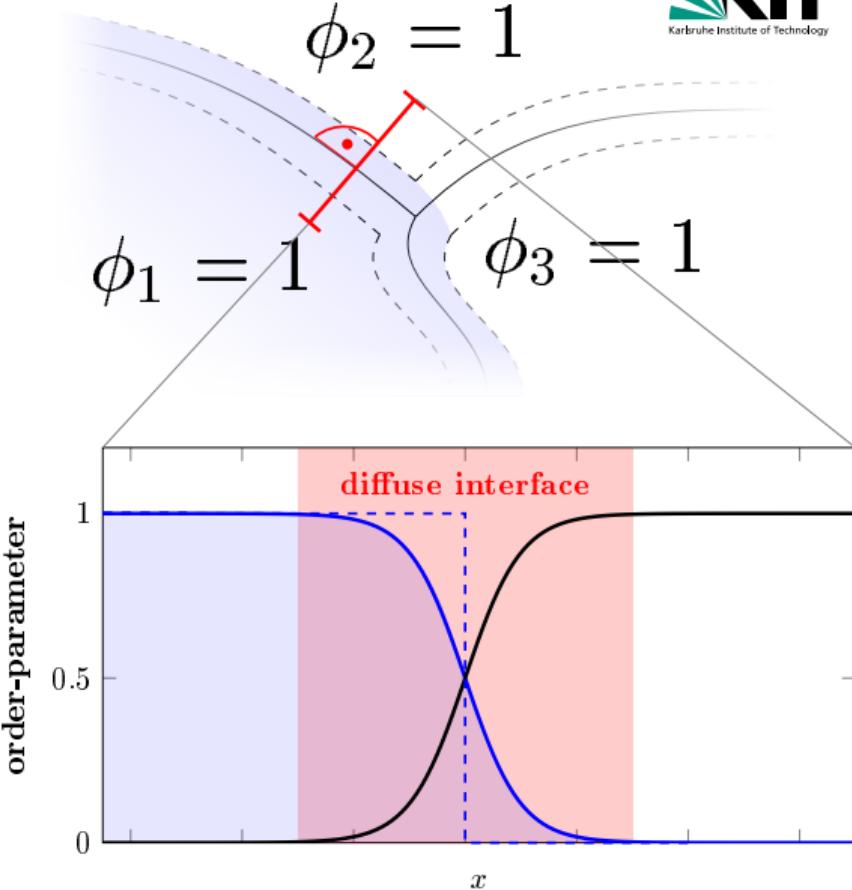
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Phase-field model for solid state sintering

- **phase-field model** based on grand potential approach
 - evolution of phase-fields $\partial_t \phi_\alpha$
 - evolution of chemical potentials $\partial_t \mu$
 - constant temperature
- different **diffusion mechanisms**:
surface (5), grain boundary (2) and
volume diffusion (1,3,4,6)
- arbitrary number **phases/grains**
- optimized on different levels
- massive-parallel PACE3D framework
- ⇒ efficient simulation of large domains

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Hötzer et al. Phase-field simulation of solid state sintering, Acta Materialia (2018)

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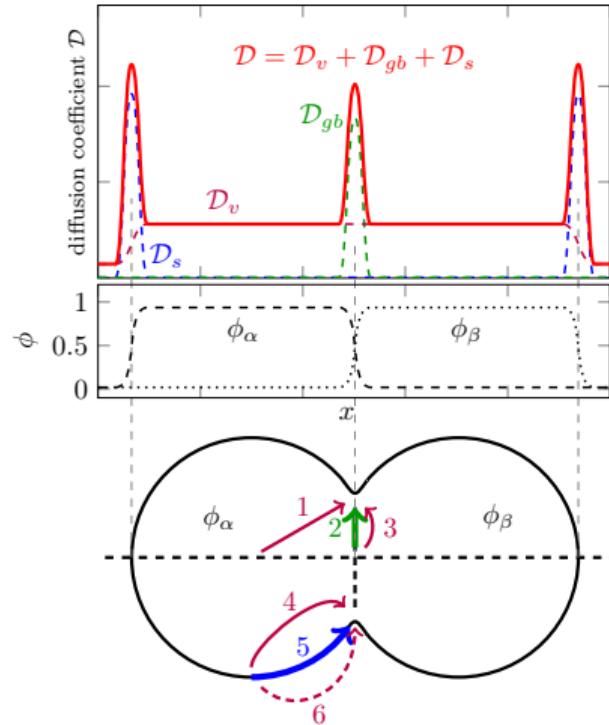
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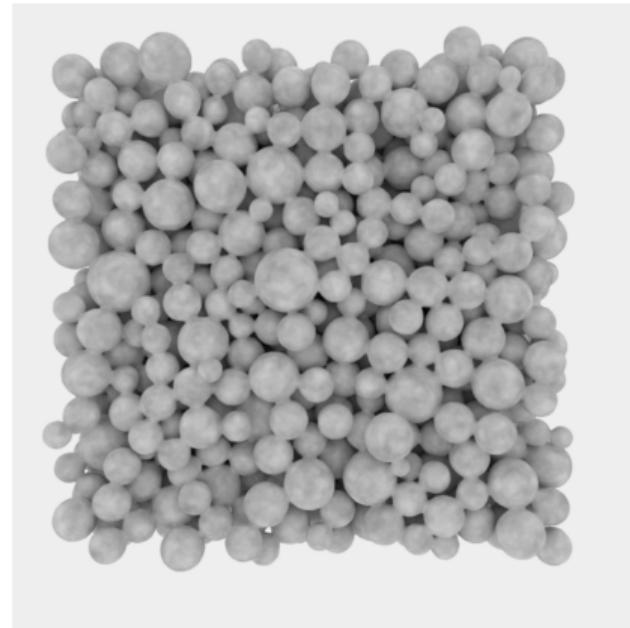
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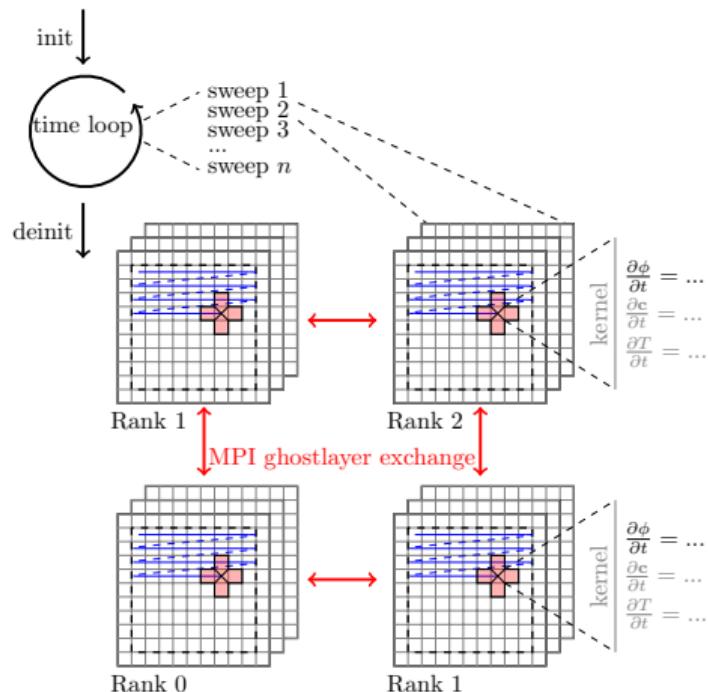
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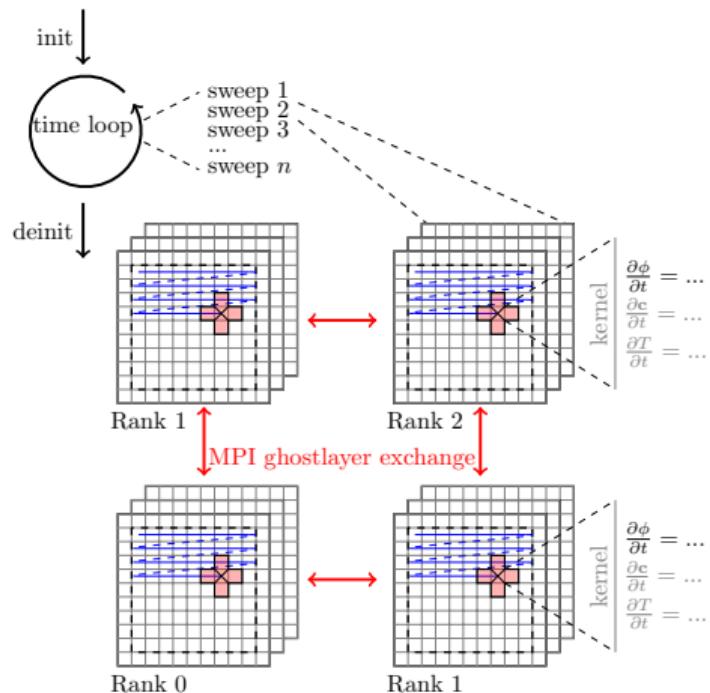
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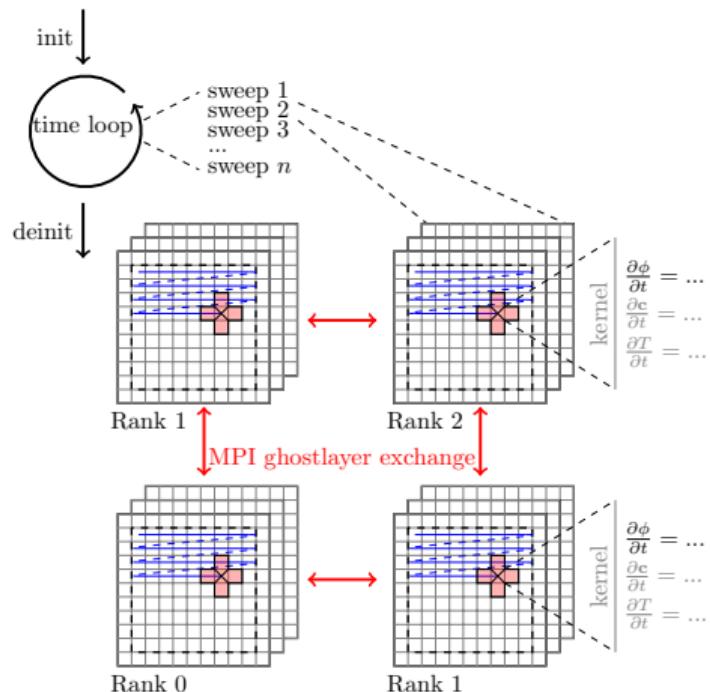
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Part IV: Implementation of the phase-field model

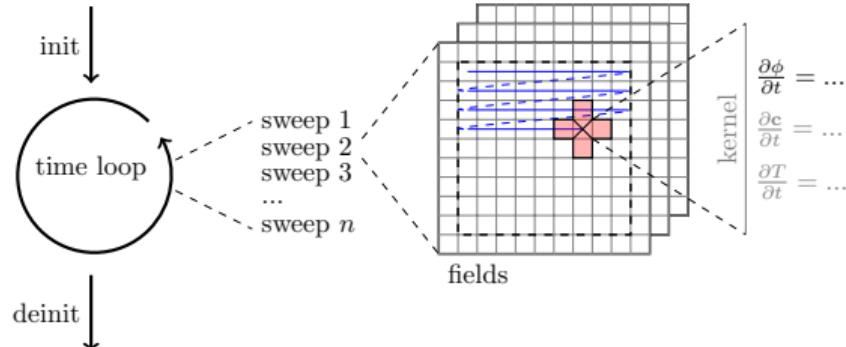
<https://www.flickr.com/photos/149836801@N02/35401845633/>

Phase-field algorithm

- lattice fields
 - two AoS for phase-field (ϕ_{src}, ϕ_{dst})
 - two SoA for chemical potential (μ_{src}, μ_{dst})
 - communication hiding for both sweeps
- storing new values calculated from *src* in *dst*

Algorithm 1 time step with overlapping communication

```
1: END: COMMUNICATION ( $\phi_{src}$ )
2:    $\phi_{dst} \leftarrow \phi\text{-sweep}\left(\phi_{src}, \mu_{src}\right)$ 
3:    $\phi_{dst}$ -BOUNDARY CONDITIONS
4: START: COMMUNICATION ( $\phi_{dst}$ )
5: END: COMMUNICATION ( $\mu_{src}$ )
6:    $\mu_{dst} \leftarrow \mu\text{-sweep}\left(\mu_{src}, \phi_{src}, \phi_{dst}\right)$ 
7:    $\mu_{dst}$ -BOUNDARY CONDITIONS
8: START: COMMUNICATION ( $\mu_{dst}$ )
9:   SWAP  $\phi_{src} \leftrightarrow \phi_{dst}$  and  $\mu_{src} \leftrightarrow \mu_{dst}$ 
```

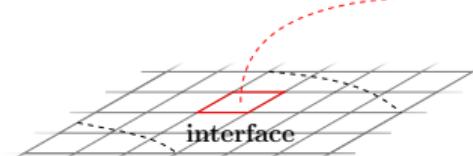


ϕ -sweep

$$\tau \varepsilon \frac{\partial \phi_\alpha}{\partial t} = -\varepsilon \underbrace{\left(\frac{\partial a(\phi, \nabla \phi)}{\partial \phi_\alpha} + \nabla \cdot \frac{\partial a(\phi, \nabla \phi)}{\partial \nabla \phi_\alpha} \right)}_{\text{D3C7}} - \underbrace{\frac{1}{\varepsilon} \frac{\partial \omega(\phi)}{\partial \phi_\alpha} - \frac{\partial \psi(\phi, \mu)}{\partial \phi_\alpha} + \lambda}_{\text{D3C1}}$$

- finite differences scheme for space
- forward Euler scheme for the time discretization
- local reduction of order-parameters with size 8
⇒ handle arbitrary number of evolution equations at cost of using lookup table for values

$$\begin{array}{ccc} \phi(\mathbf{x}, t) & \xrightarrow{\text{D3C7}} & \phi(\mathbf{x}, t + \Delta t) \\ & \searrow \text{D3C1} & \end{array}$$



index	value	e.g.
$N - 1$	ϕ_{N-1}	0
$N - 2$	ϕ_{N-2}	0
$N - 3$	ϕ_{N-3}	0.57
\vdots	\vdots	\vdots
2	ϕ_2	0
1	ϕ_1	0.43
0	ϕ_0	0

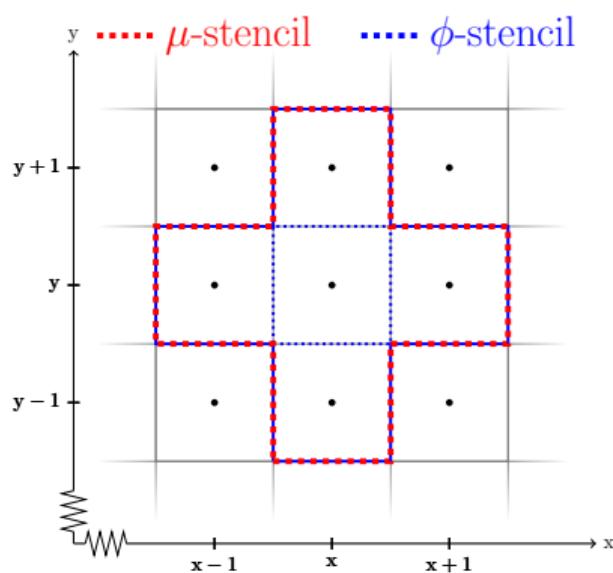
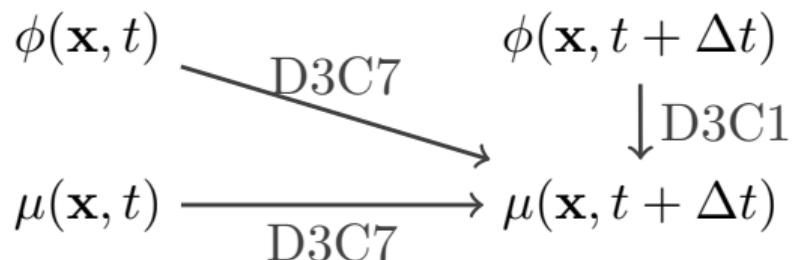


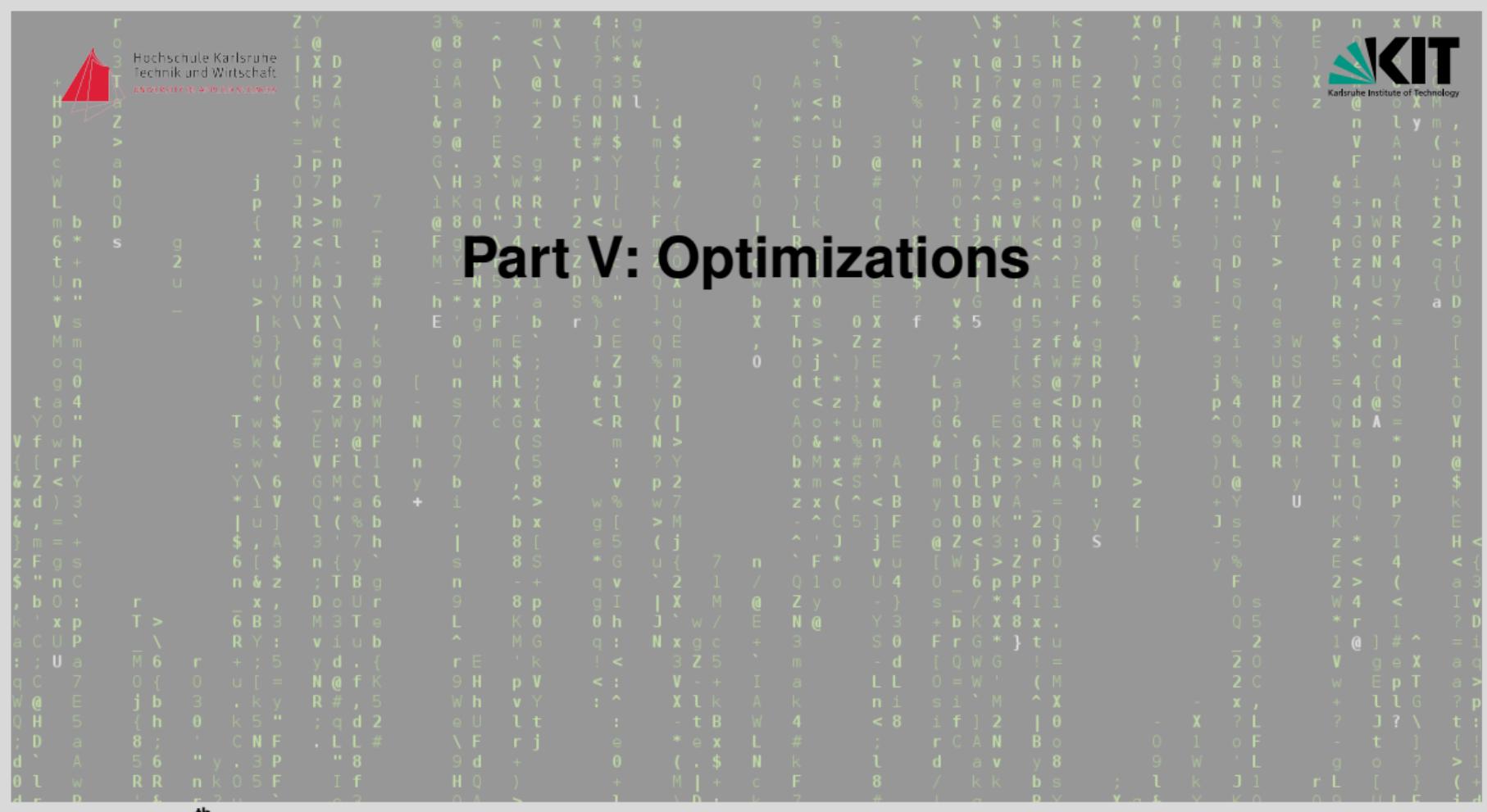
index	value	e.g.
-	-	-
-	-	-
-	-	-
-	-	-
-	-	-
$N - 3$	ϕ_{N-3}	0.57
1	ϕ_1	0.43
1	1	0.43

μ -sweep

$$\frac{\partial \boldsymbol{\mu}}{\partial t} = \underbrace{\left[\sum_{\alpha=1}^N h_\alpha(\vec{\phi}) \left(\frac{\partial \vec{c}^\alpha(\boldsymbol{\mu})}{\partial \boldsymbol{\mu}} \right) \right]^{-1}}_{\text{D3C1}} \underbrace{\left(\nabla \cdot (\mathbf{M}(\vec{\phi}, \boldsymbol{\mu}) \nabla \boldsymbol{\mu}) - \sum_{\alpha=1}^N \vec{c}^\alpha(\boldsymbol{\mu}) \frac{\partial h_\alpha(\phi)}{\partial t} \right)}_{\text{D3C7}}$$

- finite differences scheme for space
- forward Euler and implizit scheme for the time discretization





Optimizations

Parameter layer

- fitted parabolic Gibbs energies
- reduction of the parameter matrices to a class based concept

Model layer

- classification of cells
⇒ skip terms
- pre-calculation of values
- scalar mobility function instead of tensor mobility

Numeric layer

- $\partial_t \phi \Rightarrow$ forward Euler scheme
- $\partial_t \mu \Rightarrow$ forward Euler scheme, Jacobi, PCG

Algorithm layer

- streaming in sweeps
- advanced buffering techniques
- domain decomposition (MPI)
- communication hiding
- local reduction of order parameter (LROP) of the N element ϕ -vector to 8 elements
- MPI-IO and reduced mesh output

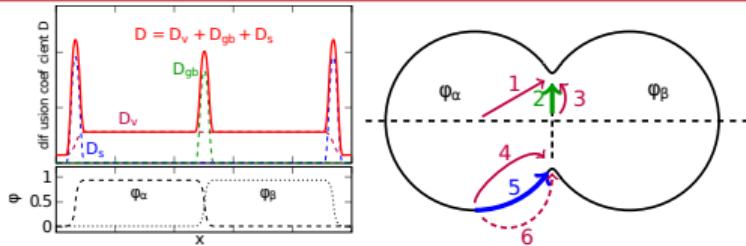
Hardware layer

- explicit vectorization of kernels
- optimized memory layouts (AoS vs. SoA)

Optimizations

Numerical layer

- diffusion coefficient differs over multiple order of magnitude
- ϕ -equ. operates on larger stable time step width as μ -equ.
- μ -equ. is relatively localized and well-conditioned
 - ⇒ different implicit scheme to solve μ -equ.
 - Jacobi (JAC)
 - Jacobi with fixed number of iterations (JACFIXN)
 - preconditioned conjugated gradient method (PCG)
 - ⇒ explicit forward euler scheme to solve ϕ -equ.

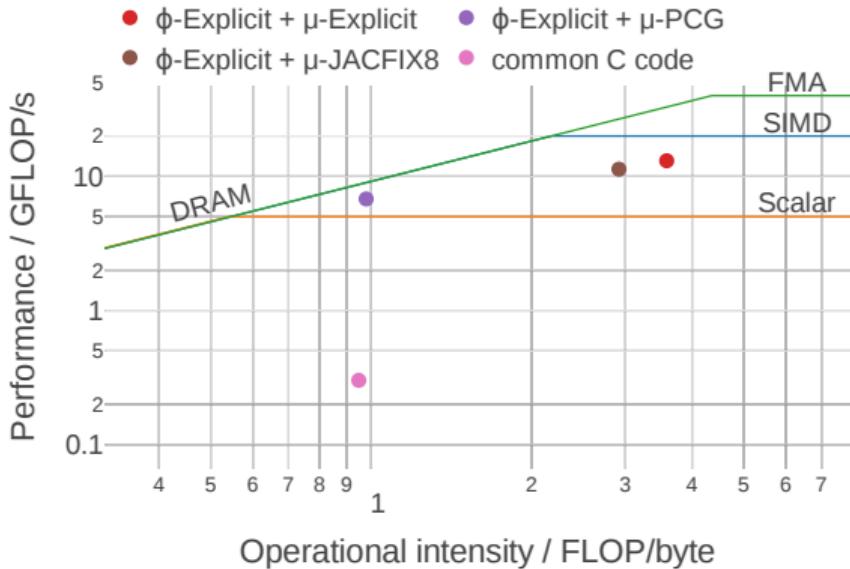




Part VI: Performance results

<https://www.flickr.com/photos/135366503N05/27140651174/>

Roofline

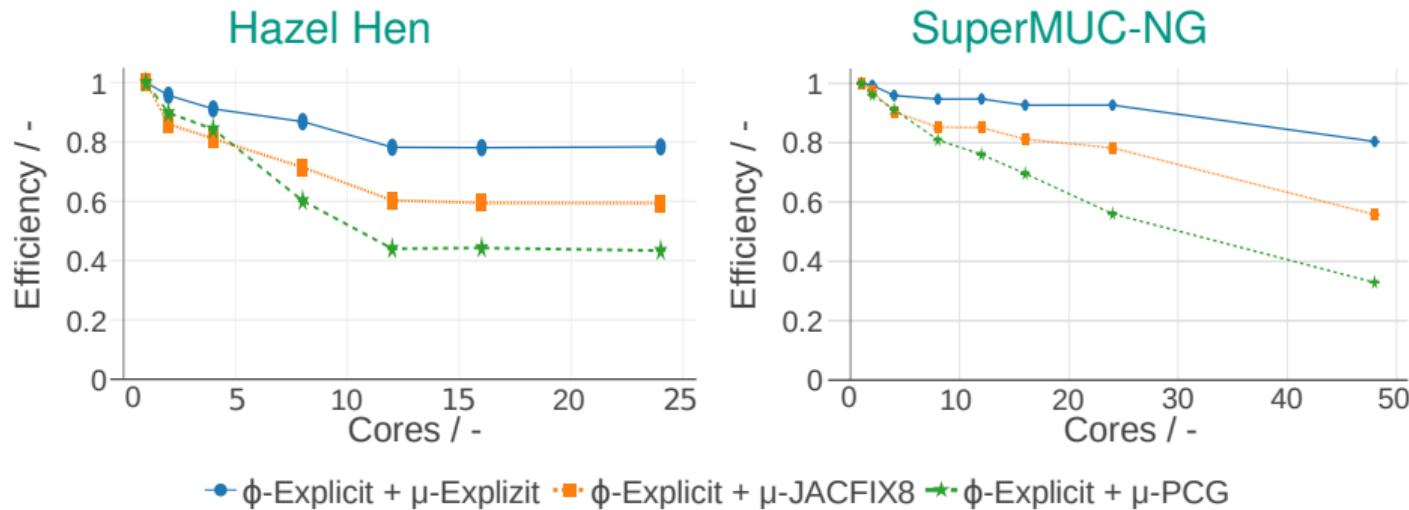


- PCG 646 times and JACFIX8 1736 faster than the common explicit C-code
- ϕ -Explicit + μ -Explicit \Rightarrow 32.5 %
- ϕ -Explicit + μ -JACFIX8 \Rightarrow 28.2 %
- ϕ -Explicit + μ -PCG \Rightarrow 16.9 %

- Hazel Hen - 2x 12 core Intel Xeon CPU E5-2680 v3
- frequency pinned to 2.5 Ghz
- single core result
- 80^3 voxel cell

Hierl, Hözter, Seiz et al. Extreme Scale Phase-Field Simulations of Sintering Processes, IEEE/ACM scA (2019)

Single node scaling

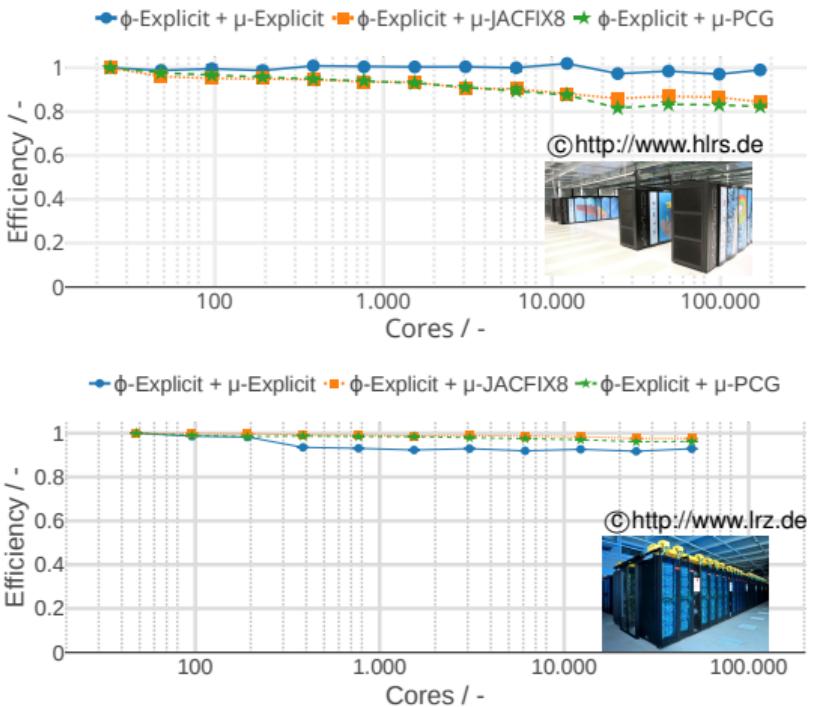


- drop due to saturation of memory bandwidth
- implicit scheme scale worse
- time to solution for implicit scheme still much smaller
 - ⇒ further cache and memory optimizations needed

- Hazel Hen - 2x 12 core Intel Xeon CPU E5-2680 v3
- SuperMUC-NG - 2x 24 core Intel Skylake Xeon Platinum 8174
- 80^3 voxel cell per rank
- frequency pinned to 2.5 Ghz
- $\Delta t_\phi = 40\Delta t_\mu$

Weak scaling results

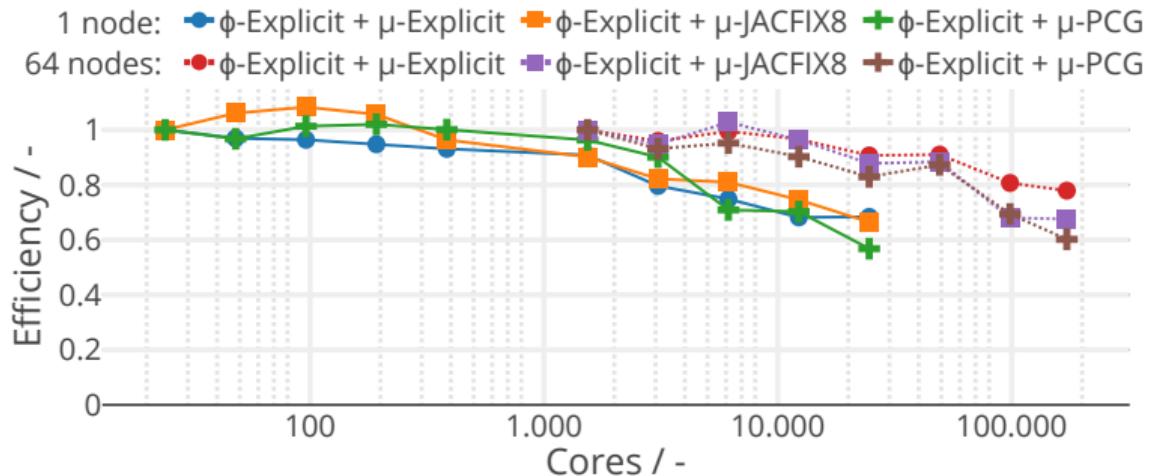
Hazel Hen
SuperMUC-NG



- proper weak scaling behavior up to 170 000 cores for all schemes

- Hazel Hen - 2x 12 core Intel Xeon CPU E5-2680 v3
- 185 088 cores \Rightarrow 7.4 PFlops
- SuperMUC-NG - 2x 24 core Intel Skylake Xeon Platinum 8174
- 304 128 core \Rightarrow 26.3 PFlops
- 80^3 voxel cell per rank
- $\Delta t_\phi = 40\Delta t_\mu$

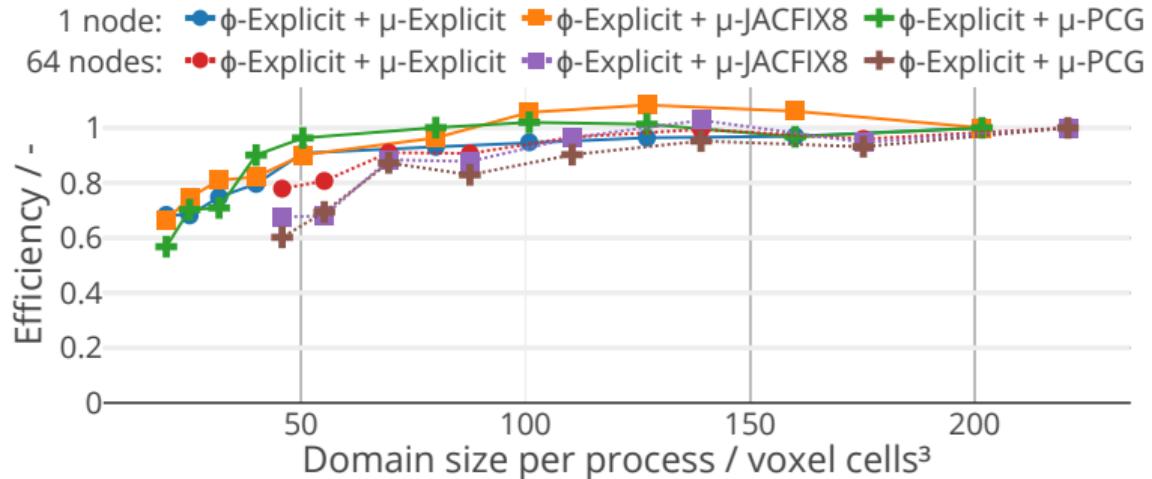
Strong scaling results



- for $> 50^3$ voxel cell domain per process:
⇒ efficiency $> 70\%$ compared to single node
- abnormal scaling behavior probably due to domain decomposition effect

- Hazel Hen - 2x 12 core Intel Xeon CPU E5-2680 v3
- 185 088 cores ⇒ peak performance: 7420 TFlops
- two settings (**1N** and **64N**), memory almost entirely used
 - 1N** starting from 1 node with $640 \times 480 \times 640$ voxel cells
 - 64N** starting from 64 nodes with $1920 \times 6720 \times 1280$ voxel cells
- $\Delta t_\phi = 40\Delta t_\mu$

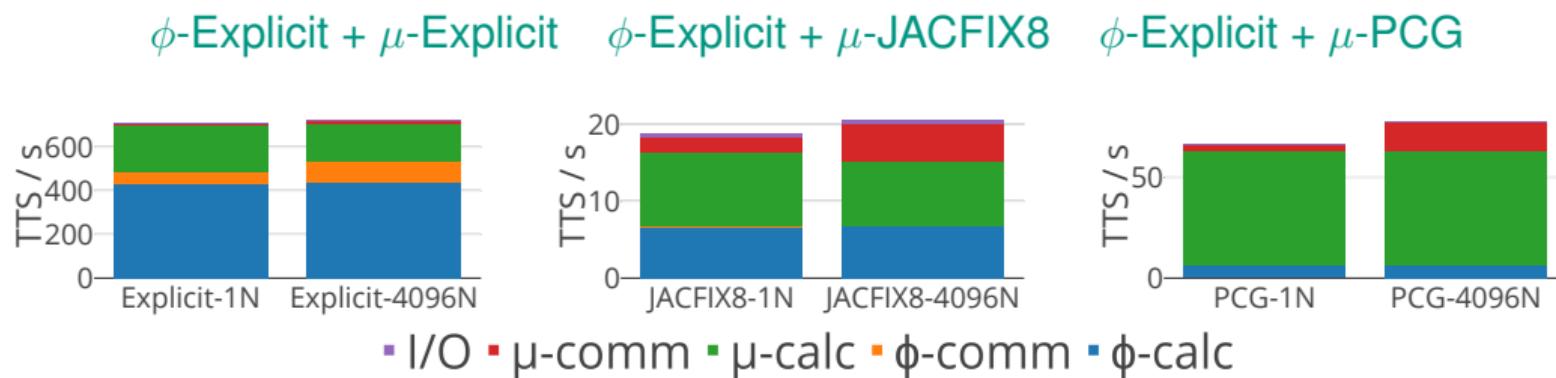
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Time to solution (TTS) comparison – Hazel Hen



- implicit schemes are much faster as explicit ones
- slight increase for JACFIX8 and PCG in TTS
- μ -communication hidden in explicit case
- ϕ -communication hidden in implicit cases
- in implicit cases μ -communication increases

- Hazel Hen - 2x 12 core Intel Xeon CPU E5-2680 v3
- comparison between one 1 node (*-1N) and 4096 nodes (*-4096N)
- $\Delta t_\phi = 40 \Delta t_\mu$

I/O optimizations

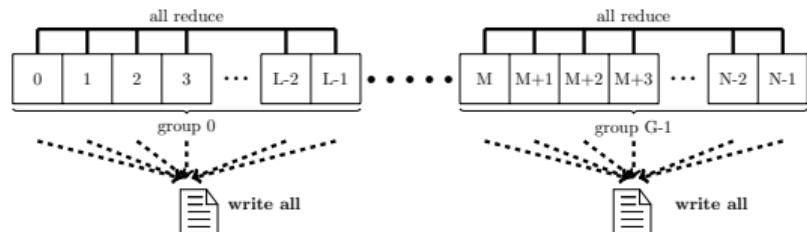
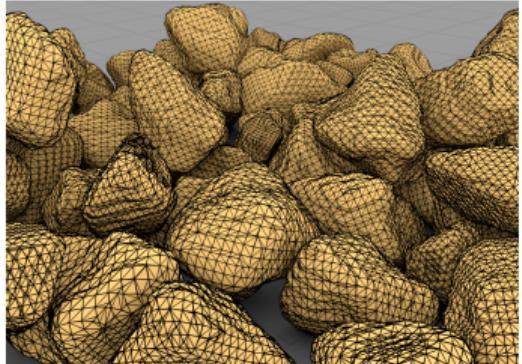
- MPI-I/O to store voxel fields for
 - detailed analysis
 - checkpointing

⇒ ~ 90 GB/s ↔ 44.5 % peak performance^a

- reduced mesh output based on STL format for
 - analysis
 - visualization

⇒ ~ 142 GB/s ↔ 71 % peak performance^b

⇒ ~ MB for STL instead of
~ GB for voxel fields



^a 2 400³ voxel = 823GB, 6 144 cores, 1728 writer on 54 OST using HLRS WS9 file system

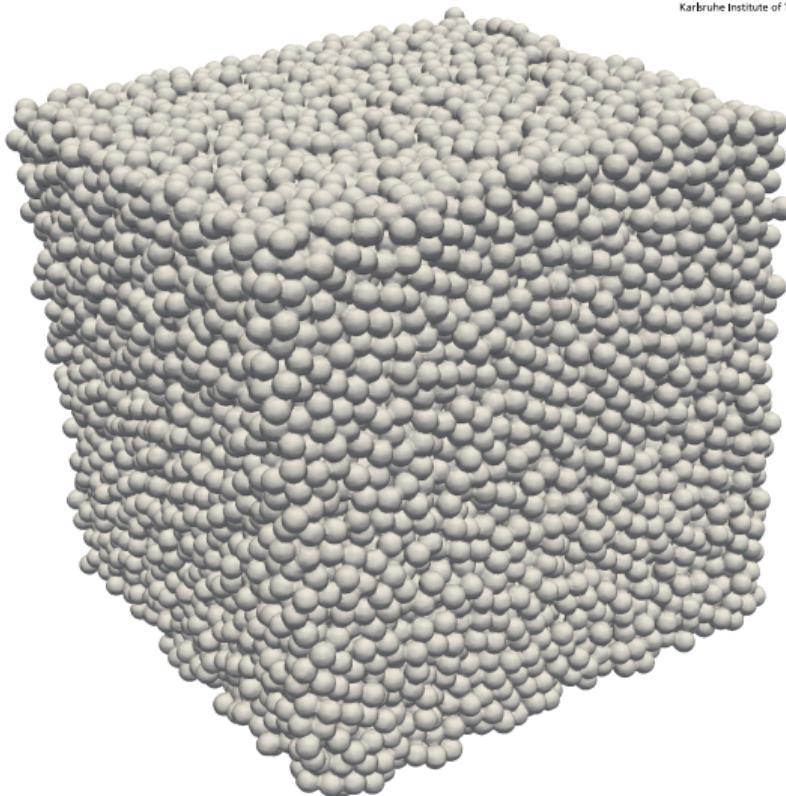
^b 2 400³ voxel, 6 144 cores, 54 MPI groups writing 193GB on HLRS WS9 file system in 3.5s including marching



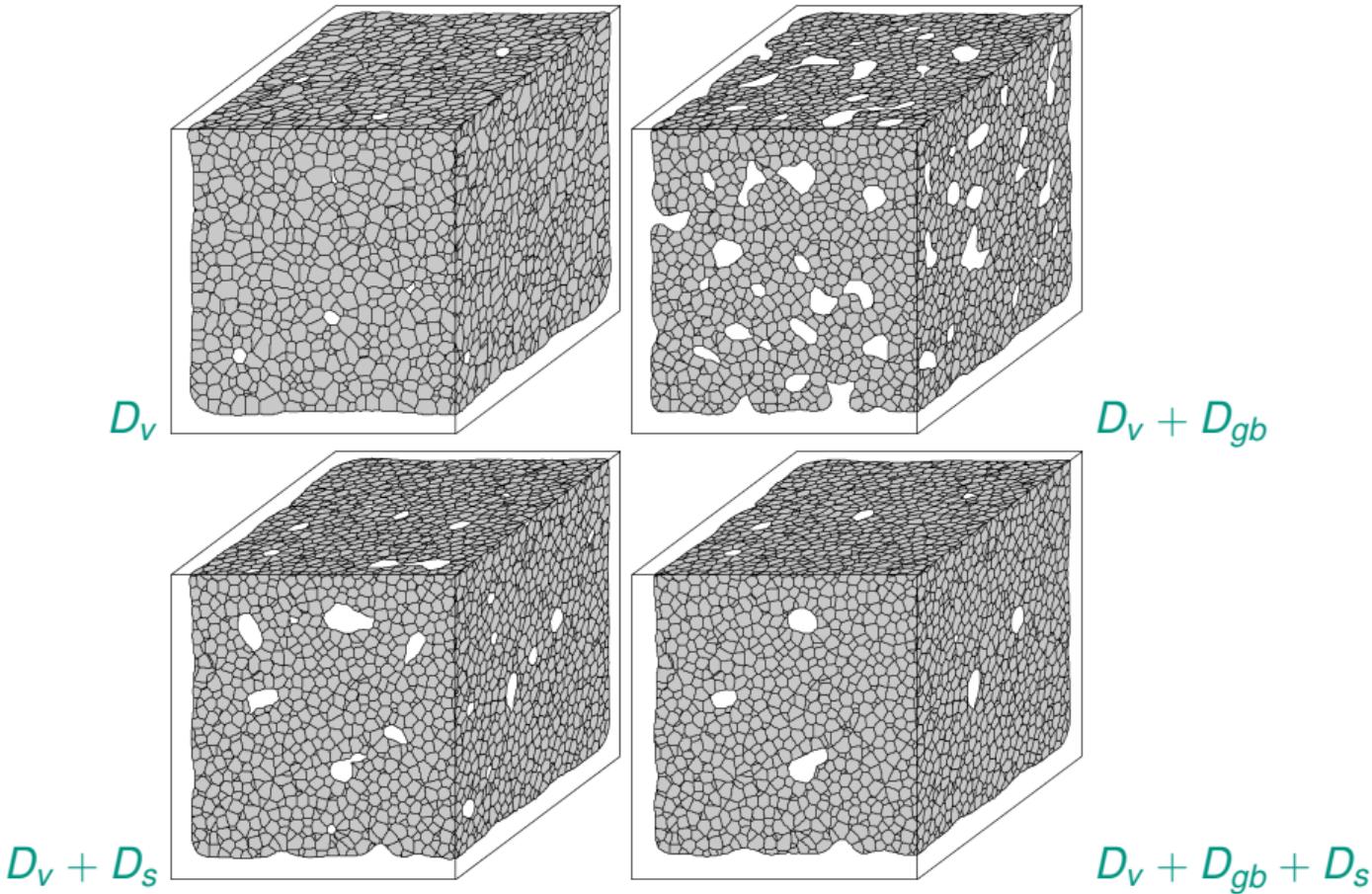
Part VII: Simulation results of the sintering process

Setting for 3D simulations

- 3D simulations based on parameters of Al_2O_3
- $400 \times 400 \times 400$ voxel cell domain
- 24 897 particles with a radius of $0.35 \mu m$
- initial density 63.8 %
- calculations on 1 083 cores of Hazel Hen supercomputer for 50h
- 10 million time steps
- different active diffusion mechanisms
 - D_v
 - $D_v + D_{gb}$
 - $D_v + D_s$
 - $D_v + D_{gb} + D_s$
- note: grain growth artificially inhibited for comparison with Coble model

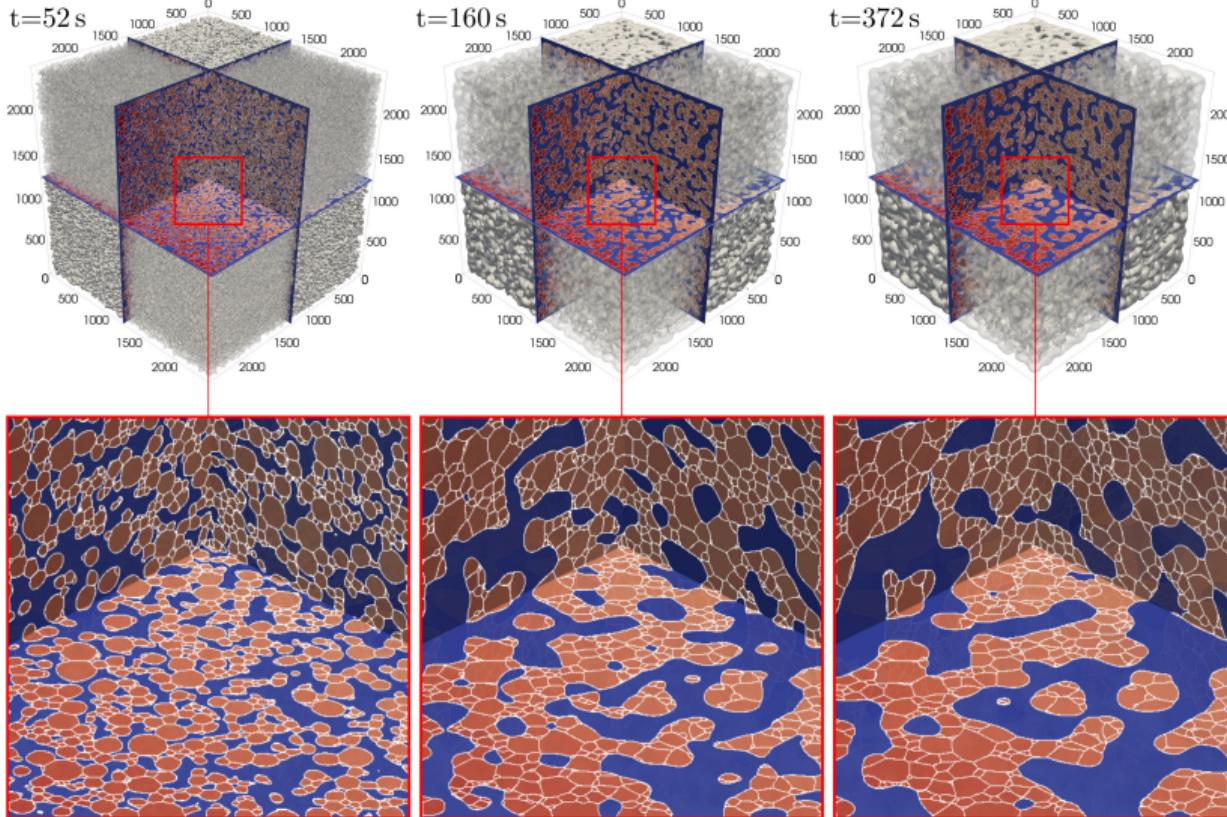


Hötzer et al. Phase-field simulation of solid state sintering, Acta Materialia (2018)



Hötzer et al. Phase-field simulation of solid state sintering, Acta Materialia (2018)

Extreme scale simulation



- parameters based on Al_2O_3
- with $D_s = D_{gb} = 1000 \cdot D_V$
- $2400 \times 2400 \times 2400$ voxel cells
 $\Rightarrow 120 \times 120 \times 120\mu\text{m}$
- initially packed with 1191901 individual particles
- computed with 168 000 cores of the Hazel Hen supercomputer

Hierl, Hözter, Seiz et al. Extreme Scale Phase-Field Simulations of Sintering Processes, IEEE/ACM scA (2019)

Conclusions

- synergistic optimizations provided a 31-fold improvement for the explicit scheme
- implicit schemes 620 (PCG) and 1 736 (JACFIX8) faster than common C code
- up to 32.5 % single-core peak performance
- mostly ideal weak-scaling behavior of explicit and implicit schemes in multi node case
- up to 24.5 % peak performance on 98 304 cores
- ability to:
 - solve the phase-field equation for millions of unknowns
 - efficiently calculate huge domains
- both of which are necessary to resolve realistic particle size distributions over multiple orders of magnitude (nm— μ m)
- allows improving the understanding of sintering materials as well as the formation of glaciers

⇒ **wholistic optimizations and parallelization enable**
the investigation of realistic microstructures

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Open questions? Ideas? Improvements?

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